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Improving Particulate Matter Source Apportionment for Health Studies: A Trained Receptor Modeling Approach with Sensitivity, Uncertainty and Spatial Analyses

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Acknowledgments

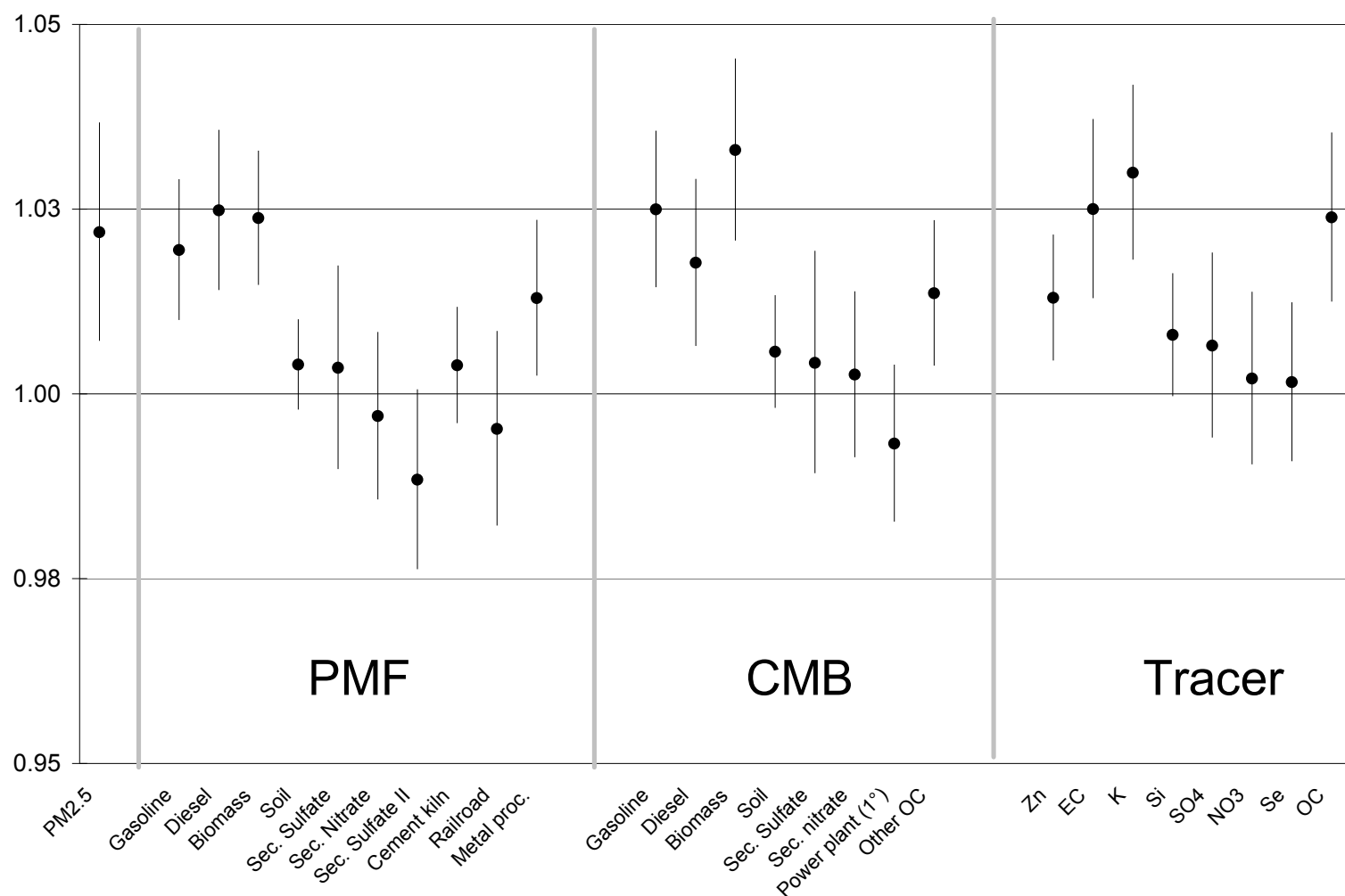
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Air Quality Models and Health Impact Assessment

- (How) Can we use "air quality models" to help identify associations between ozone PM sources and health impacts?
 - Species vs. sources
 - Very different than for traditional air quality management
 - Though this is still a very important application

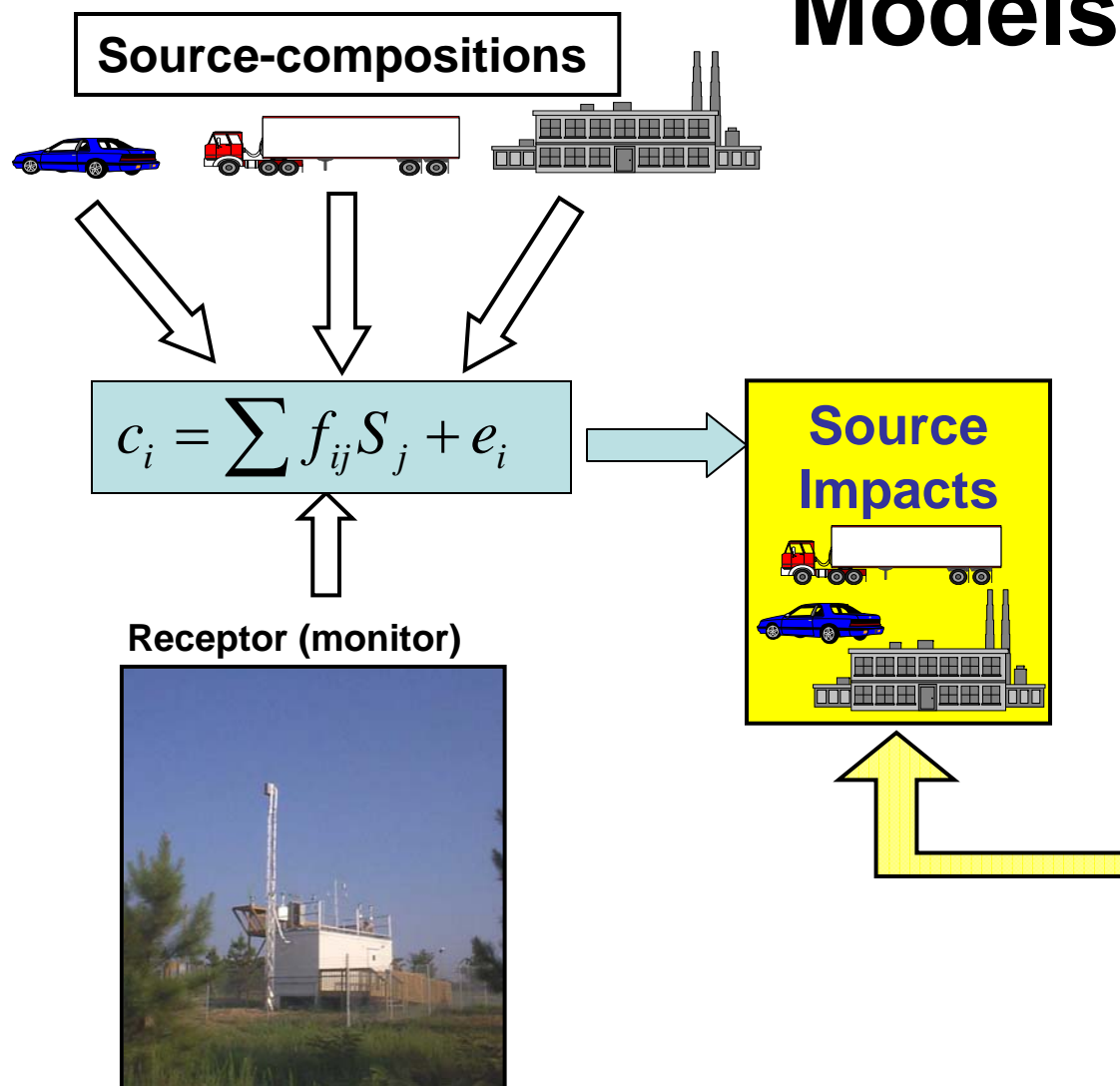


Use of Source Apportionment Results in Epidemiologic Studies

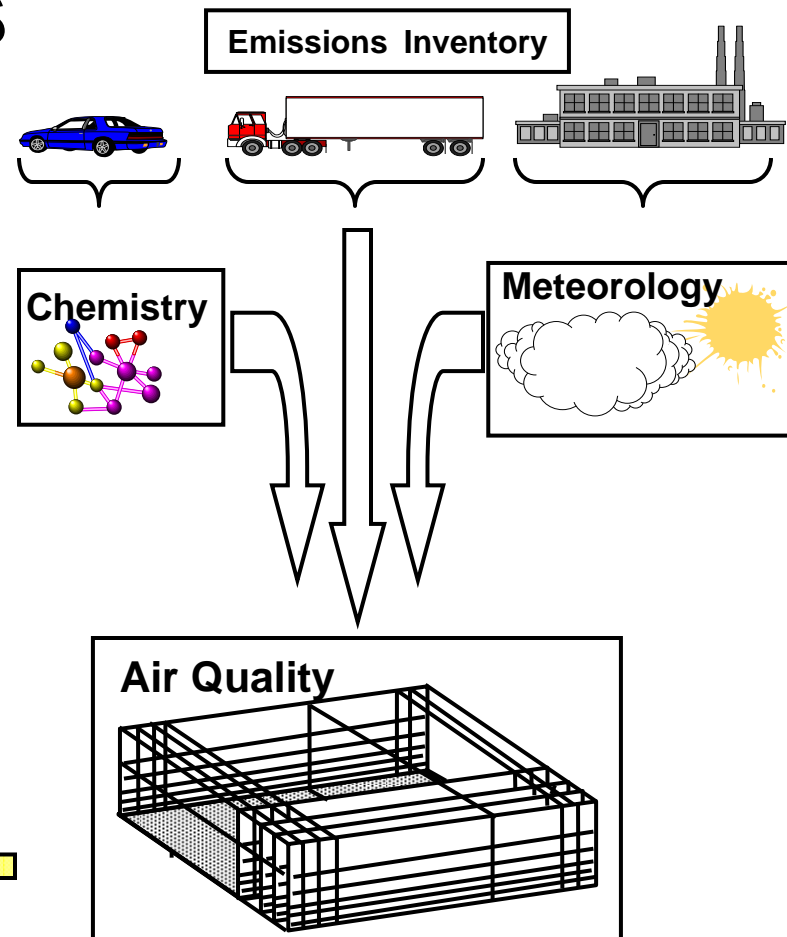


Sarnat, J. A.; et al.. *Environ. Health Perspect.* **2008**, 116, (4), 459-466.

Receptor vs. Emissions-Based Models



Receptor Model

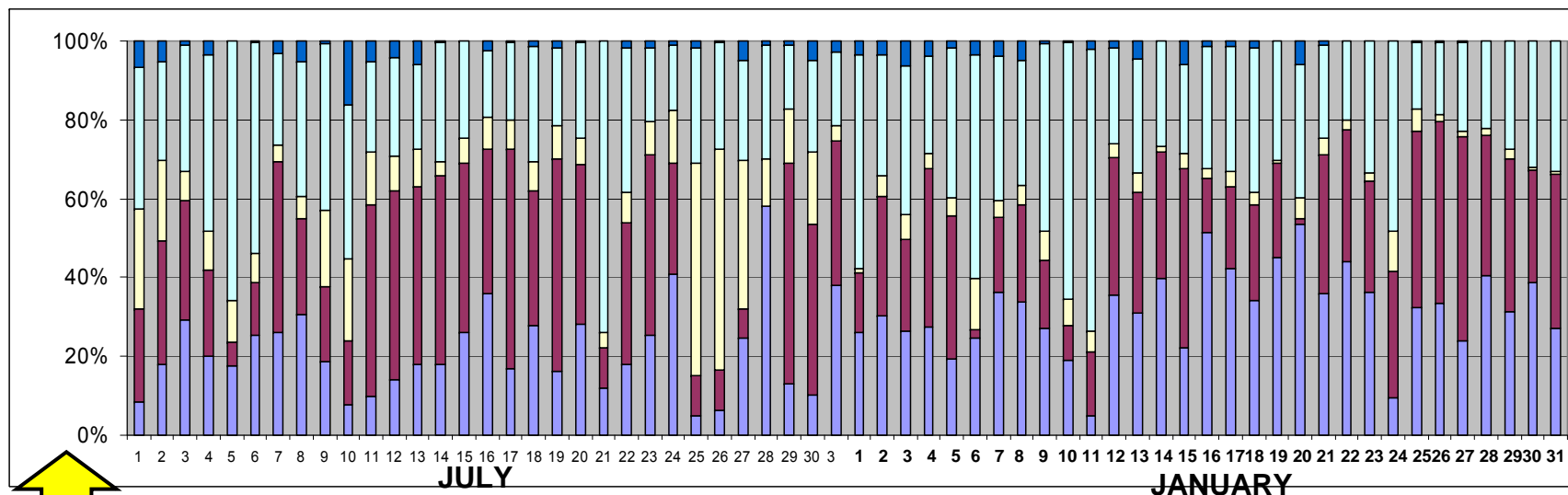


*Emissions-based Model
(3D Air-quality Model)*

The Problem!

Daily source apportionment (SA) results for Atlanta based on receptor and grid-based model results

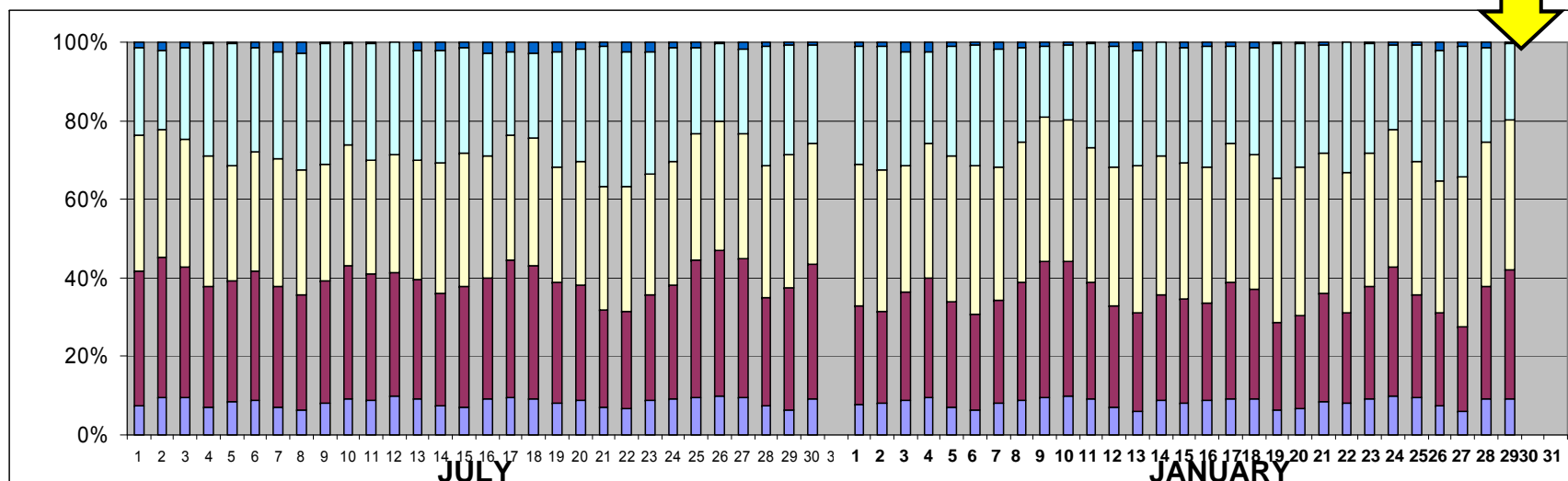
CMB



Significant daily variation

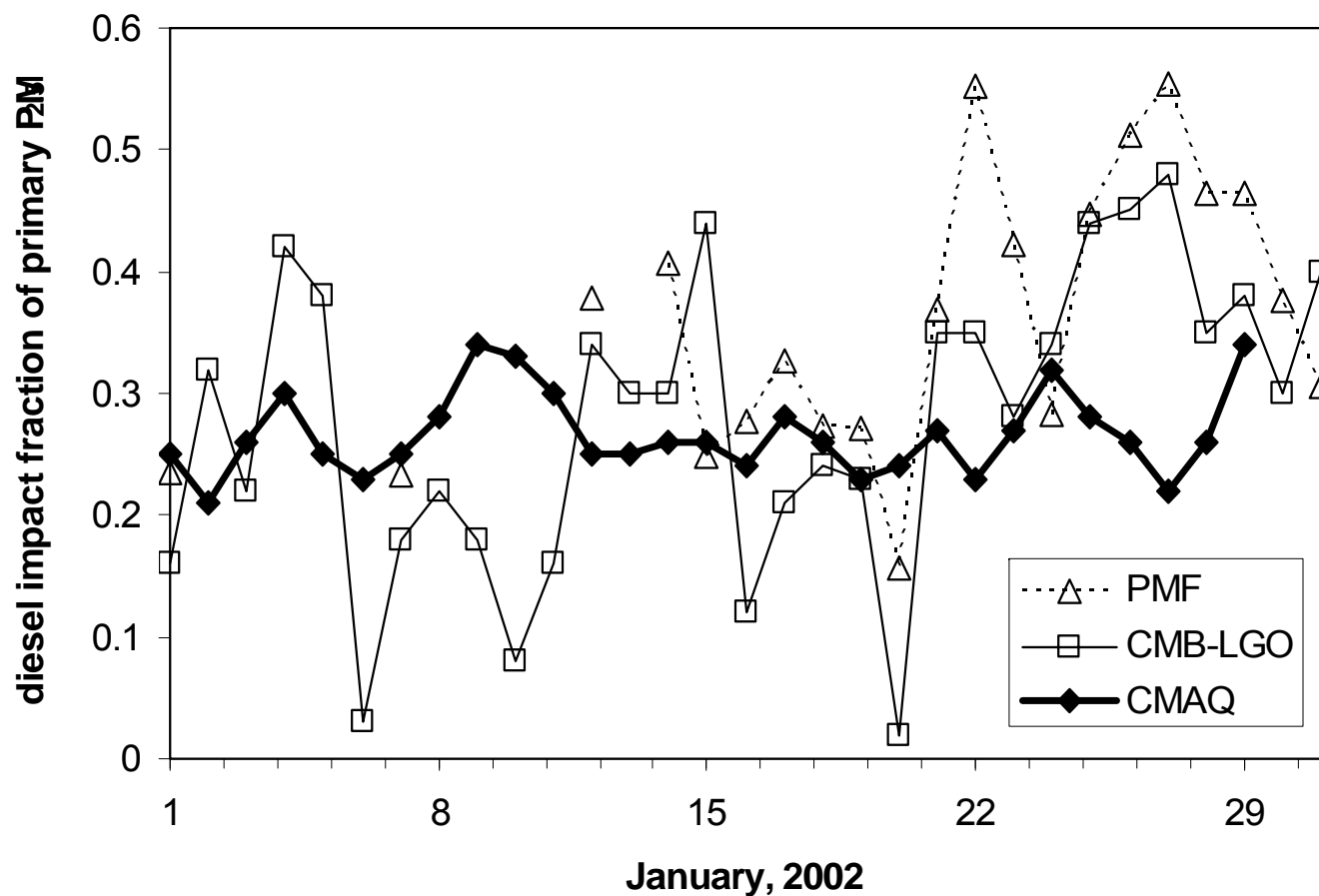
Little daily variation

CMAQ



LDGV HDDV SDUST BURN Coal

Diesel Impact Variation



Source correlations
about 0.9 for
CMAQ SA's

Both Results are Flawed

- Receptor
 - Too much day-to-day variability
 - Hard to imagine that diesel and coal burning impacts go to zero on some days and are significant on others
 - Missing sources
 - Little way around this
 - Source profiles uncertain and variable (plus that SOA issue)
- Grid-based
 - Too little variability
 - Tied to lack of small scale structure in met and emissions
 - Inconsistent with data
 - Inventories everywhere are uncertain (wrong)
- Can try to justify results
 - Our tests suggest arguments on both sides fail
- Use of source apportionment results for acute response epidemiologic analyses rely on getting day-to-day variability correct
 - Want to develop a more accurate SA for acute studies

Executive Summary

- Develop a flexible and extensible approach for source apportionment (SA)
 - Air quality management and epidemiologic studies
- Ensemble-trained approach
 - Integrate grid-based and multiple receptor modeling approaches
- Provide a tested method that directly addresses limitations in current SA methods, in particular variability, biases, and intensive resource requirements
 - Use SA results in epi studies of Atlanta and St. Louis

Limitations of Source Apportionment (SA) Approaches

- Receptor-based SA models:
 - biased estimates of primary source impacts
 - inability to identify or separate source impacts
 - excessive day-to-day variability
 - multiple zero impact days for sources that are known to be present (e.g. diesel vehicles, power plants)
 - results are representative for only the observation location
 - Some approaches resource intensive (detailed organic speciation)
- Emission-based chemical transport models (CTM):
 - large computational cost
 - results lack significant day-to-day variation in relative source impact
- Is it possible to improve results by taking an ensemble average of multiple approaches?
 - Then use ensemble results to train a receptor model

Ensemble SA and Training

- Develop SA results from weighted average of multiple methods over limited period
 - Chemical transport model (CTM)
 - Chemical mass balance models (CMB)
 - Regular (metals, ions, EC/OC)
 - Molecular Marker (MM: detailed organic speciation)
 - LGO (optimized profiles and constraints)
 - Positive matrix factorization (PMF)
 - Limited period allows using methods that are more resource intensive
 - Multiple methods allow estimating uncertainties
- Use ensemble results to develop optimized source profiles
 - Seasonally varying, location specific
- Use new profiles to calculate SA results over extended periods

Initial Application

1. **Ensemble source impacts** for July 2001 and January 2002 were developed by weighted averaging source impacts from a CTM (CMAQ) and multiple receptor-based approaches (CMB, CMB-MM, CMB-LGO, PMF).
2. **Ensemble-based source profiles (EBSPs)** for summer (July 2001) and winter (January 2002) were developed using ensemble-trained source impacts in CMB-LGO.
3. **New source impacts** were determined using CMB-LGO for a 12 month data set of daily PM_{2.5} measurements at the Atlanta, GA, Jefferson Street (JST) site using EBSPs.

Step 1: Ensemble-Trained Source Impacts

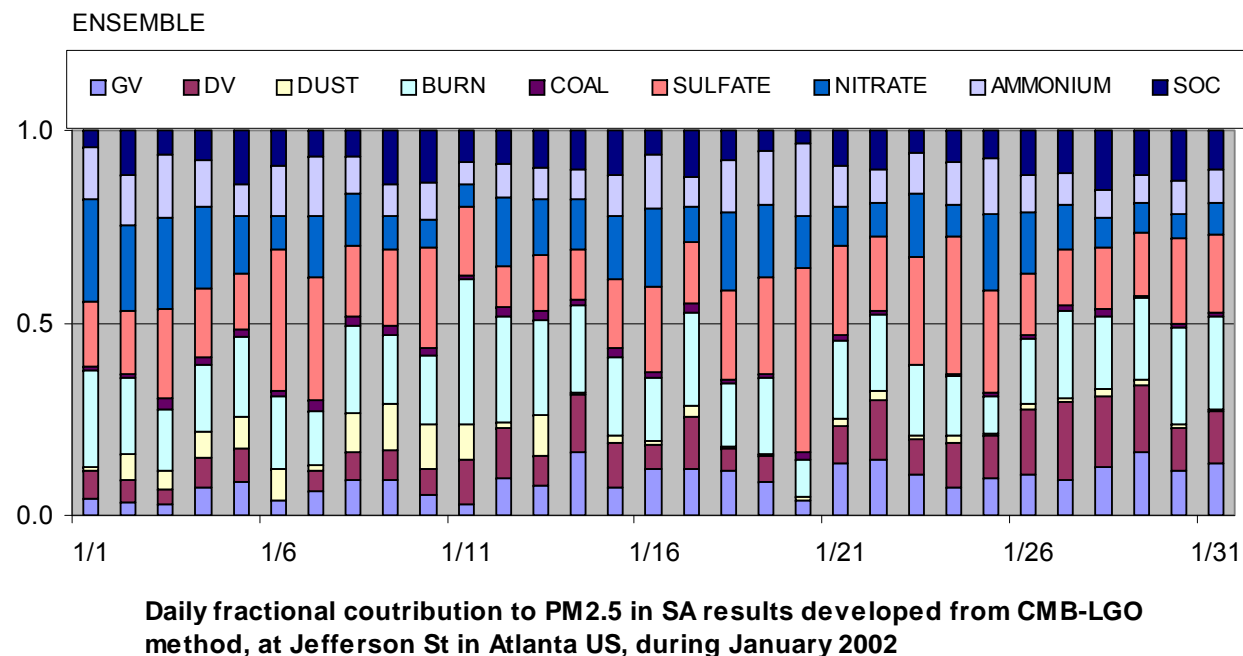
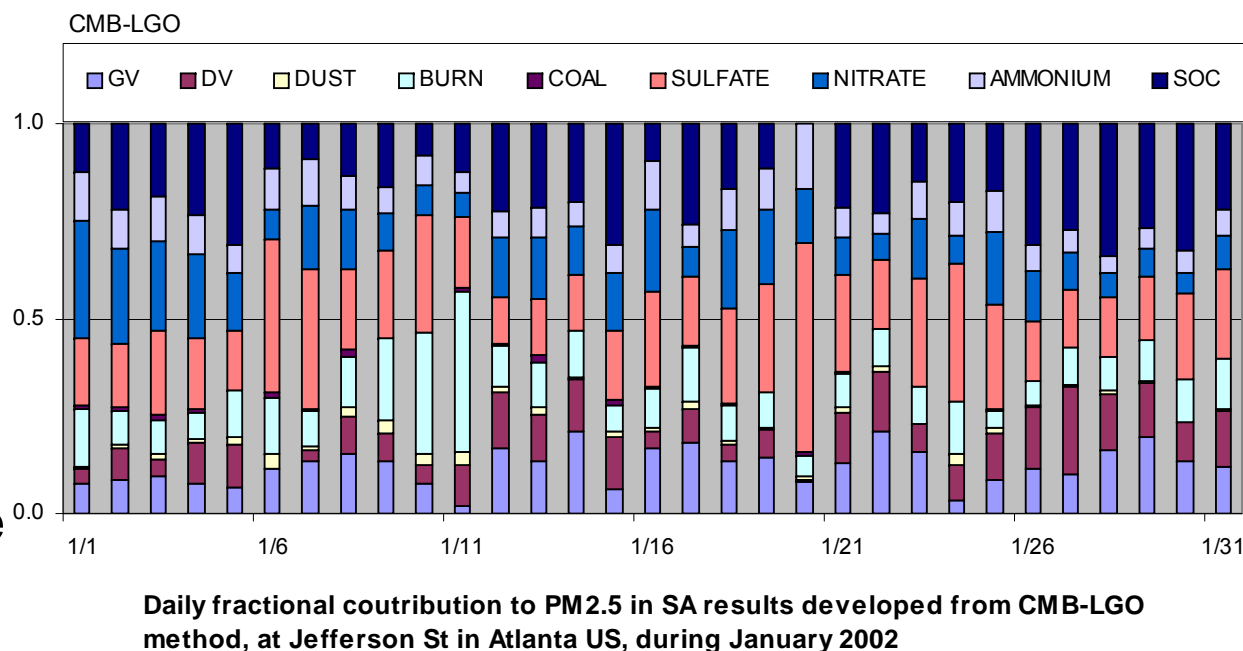
- Run L individual SA methods (CMB, CMB-MM, CMB-LGO, PMF, CMAQ) to develop weighted source impacts

$$\bar{S}_j(t_k) = \frac{\sum_{l=1}^L w_{jl}(t_k) \cdot S_{lj}(t_k)}{\sum_{l=1}^L w_{jl}(t_k)} \quad w_{jl} = \frac{1}{\sigma_{S_{lj}}^2}$$

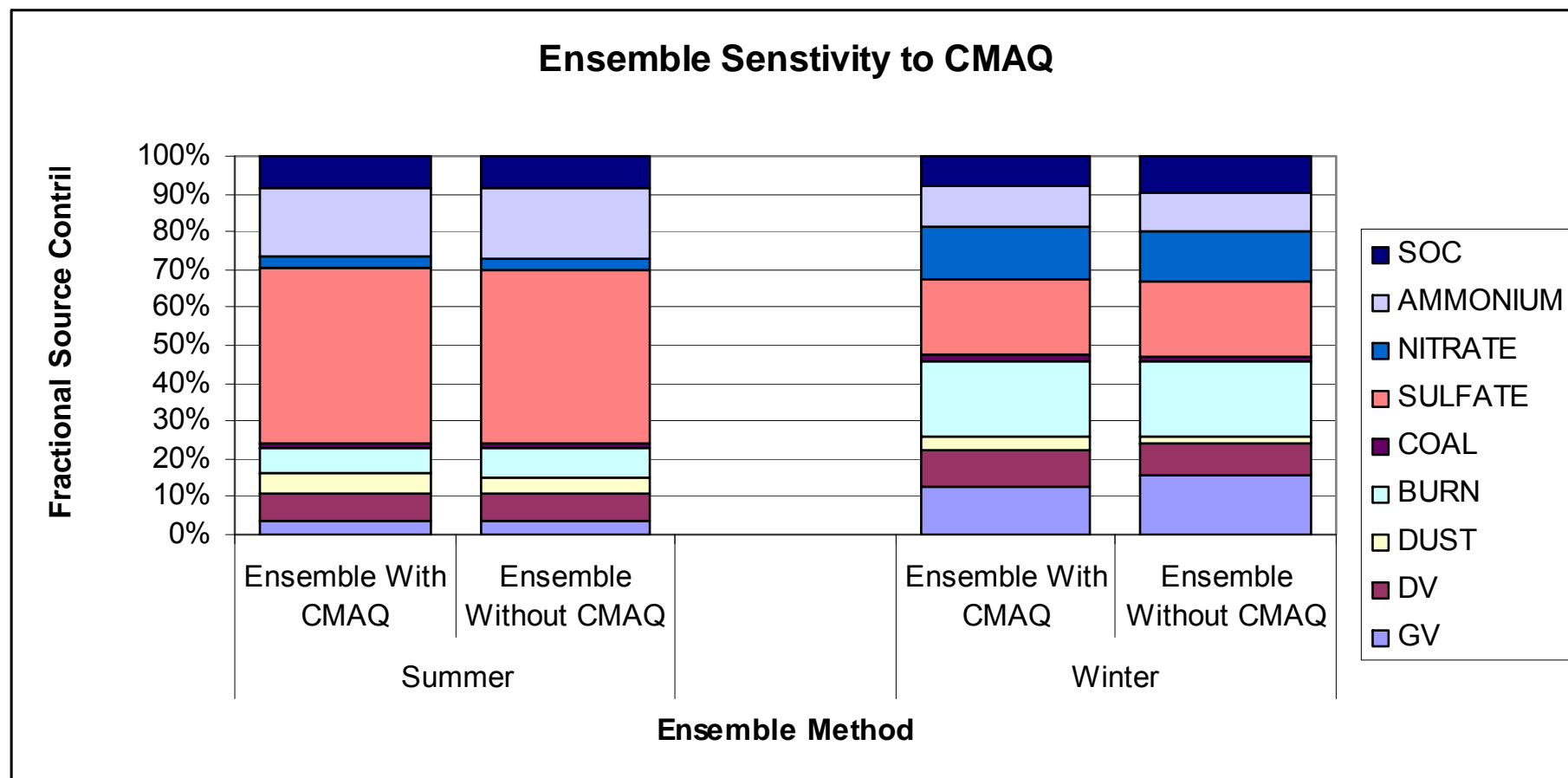
- $\bar{S}_j(t_k)$ is the ensemble-calculated impact of source j (in ug/m³) at time t_k
- $S_{lj}(t_k)$ is the impact developed by method l
- Weights, w_{jl} , are inversely proportional to uncertainty (derived from method application)

Ensemble results have less day-to-day variation in source impacts and fewer biases between observed and estimated PM_{2.5} mass compared to the original receptor model results.

Ensemble results show increases in road dust, biomass burning, and coal combustion impacts, but SOC impacts decrease.



Ensemble Sensitivity to CMAQ



Step 2: Ensemble-Based Source Profiles (EBSPs)

- Problem: won't usually have multiple methods to ensemble for large data sets (e.g. 10 yrs).
 - CMAQ and CMB-MM
- Use ensemble source impacts from small data set (e.g. 1 month) to determine ensemble based source profile.
- EBSPs (f_{ij}) were treated as the unknown in the CMB equation and solved by minimizing least squares error.

$$C_{ik} = \sum_j^J \overline{f}_{ij} \cdot \overline{S}_{jk} + e_{ik}$$

$i = \text{species}$

$j = \text{source}$

$k = \text{sample}$

$$\chi^2 = \sum_{k=1}^K \sum_{i=1}^I \frac{(C_{ik} - \sum_j^J \overline{f}_{ij} \cdot \overline{S}_{jk})^2}{\sigma_{c_{ik}}^2}$$

\overline{f}_{ij} = source profile

\overline{S}_{jk} = ensemble source impact

Step 3: New Source Impacts

- Develop new source impacts, S_{jk}^* , by minimizing least squares error using EBSPs
- CMB-LGO was run using EBSPs for a summer period (Mar - Oct) and a winter period (Jan - Feb, and Nov - Dec) and compared with measurement based source profiles (MBSPs)

$$C_{ik} = \sum_j^J \overline{f}_{ij} \cdot S_{jk}^* + e_{ik}$$

$i = \text{species}$

$j = \text{source}$

$k = \text{sample}$

$$\chi_k^2 = \sum_{i=1}^I \frac{(C_{ik} - \sum_j^J \overline{f}_{ij} \cdot S_{jk}^*)^2}{\sigma_{c_{ik}}^2}$$

$\overline{f}_{ij} = \text{EBSPs}$

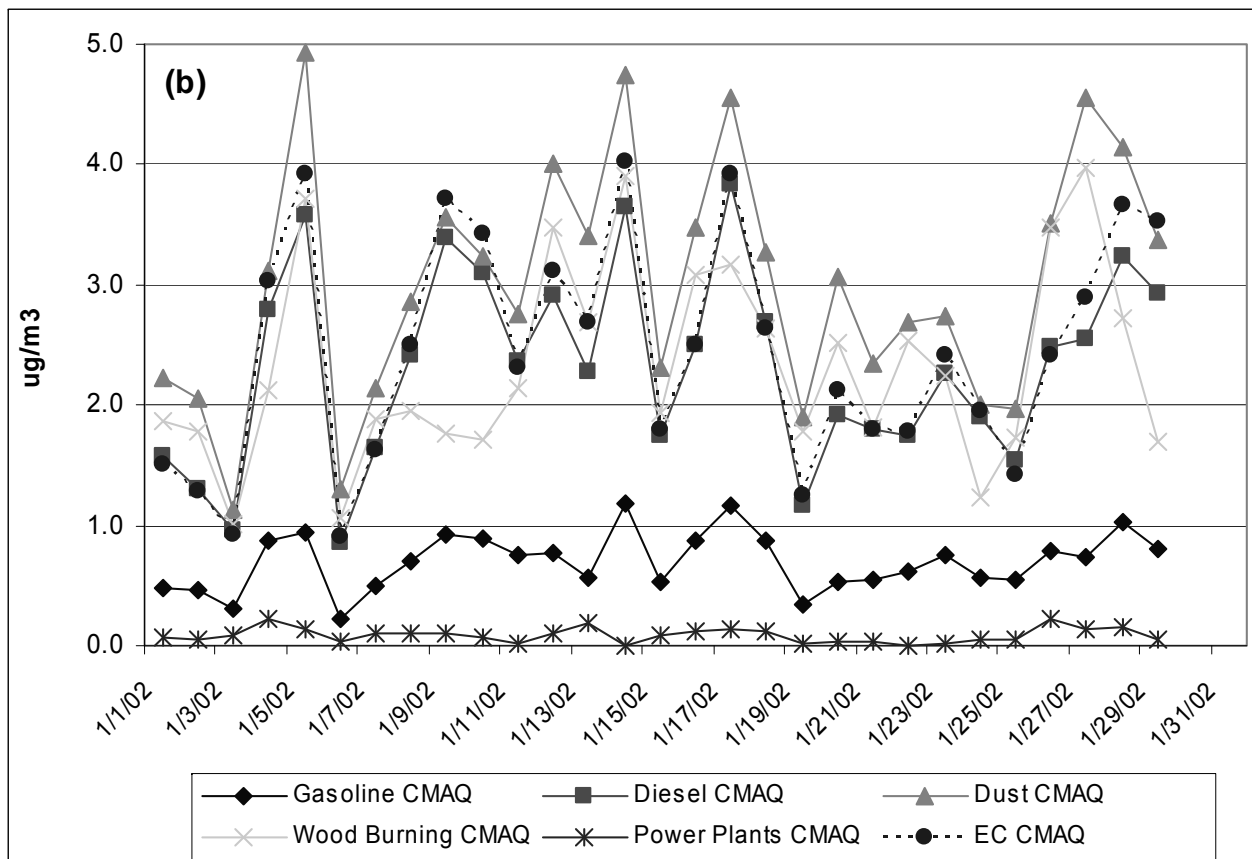
$S_{jk}^* = \text{new source impacts}$

Conclusions

- Ensemble-based source apportionment method developed to address limitations in current methods
 - Initial application to Atlanta
- Ensembling decreases variability and number of zero impact days
 - Assessed impact of including CMAQ results
- Ensemble-trained/based source profiles (EBSPs) developed for summer and winter
 - Results suggest seasonal variability in OC:EC ratios in profiles
- Application of EBSPs decreased variability, improved performance
 - Increased biomass burning and road dust impacts, decreased SOA in winter
- Future work will include
 - Applying the method to longer time periods and other locations
 - Assessing variability and refining the ensemble method
 - Using different approaches to estimating weights and assessing uncertainties
 - Conduct spatial analyses
 - Apply to more routine monitoring data
 - Incorporating source impacts into epidemiology studies

CMAQ Results

Source impacts at Jefferson Street, SEARCH site, Atlanta



Correlations between sources high ($R \sim 0.9$)